

# Sorting Problem

CSC 209 Data Structures

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# Lecture Plan

- Sorting Problem
- Insertion Sort
- Divide-and-Conquer
- Mergesort
- Quicksort
- Implementation of insertion sort, mergesort, quicksort

# Sorting Problem

**Input:** A sequence of  $n$  numbers  $\langle a_1, a_2, \dots, a_n \rangle$

**Output:** A permutation (reordering)  $\langle a'_1, a'_2, \dots, a'_n \rangle$  of the input sequence such that  $a'_1 \leq a'_2 \leq \dots \leq a'_n$

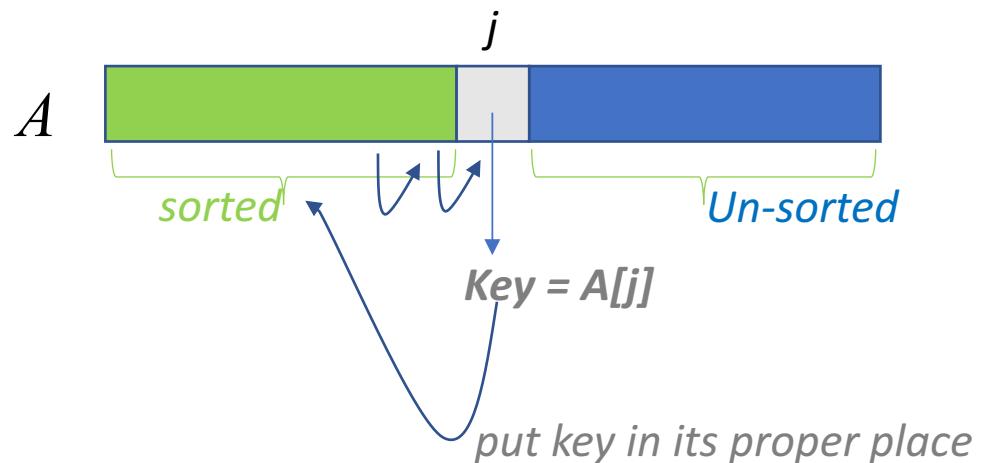
**Example:**      *Input*      8    2    4    9    3    6

*Output*    2    3    4    6    8    9

# Insertion Sort

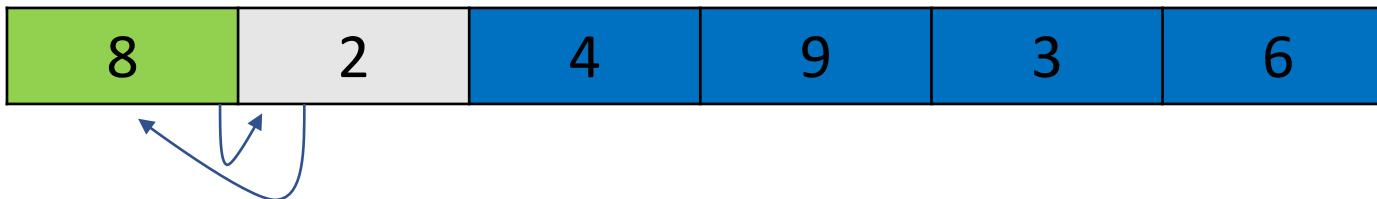
INSERTION-SORT( $A$ )

```
1  for  $j = 2$  to  $A.length$ 
2       $key = A[j]$ 
3      // Insert  $A[j]$  into the sorted sequence  $A[1 \dots j - 1]$ .
4       $i = j - 1$ 
5      while  $i > 0$  and  $A[i] > key$ 
6           $A[i + 1] = A[i]$ 
7           $i = i - 1$ 
8       $A[i + 1] = key$ 
```

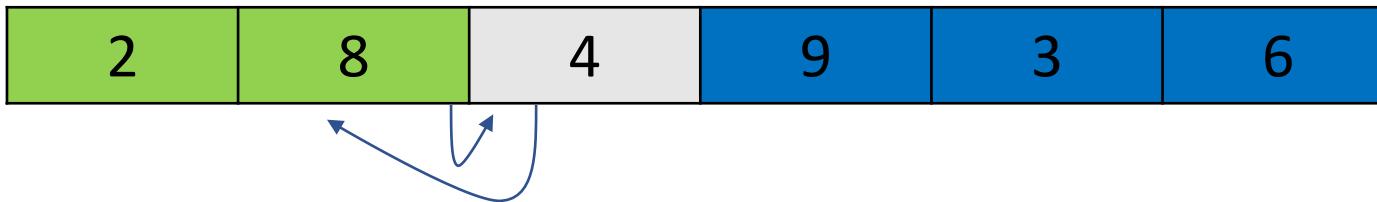


# Insertion Sort

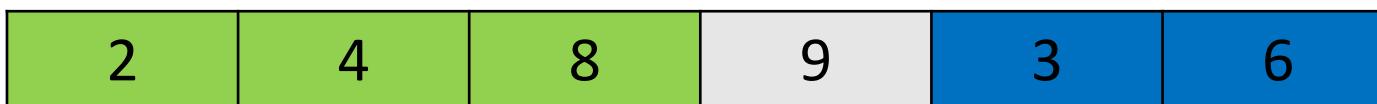
$j = 2, \text{key} = 2$



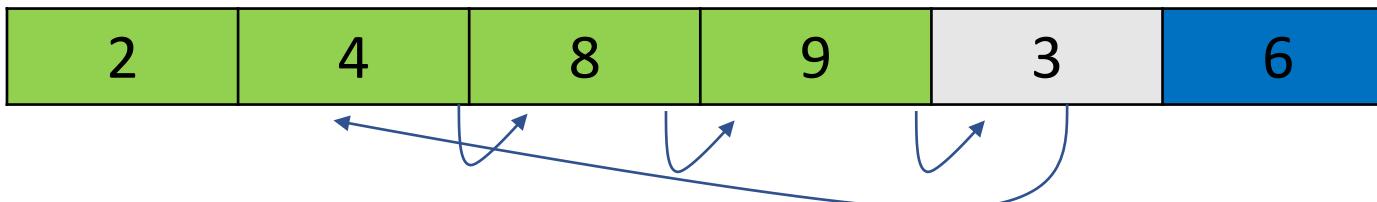
$j = 3, \text{key} = 4$



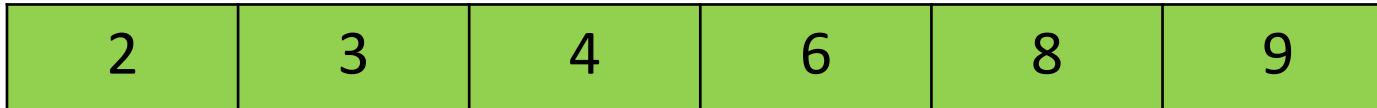
$j = 4, \text{key} = 9$



$j = 5, \text{key} = 3$



$j = 6, \text{key} = 6$



# Running Time: Insertion Sort

- Depends on the input (e.g. sorted, reverse sorted)
- Depends on the input size
- In general, we want to know the upper bound of the running time because it represents a guarantee

# Types of Running Time

- **Worst-case running time**

$T(n)$  : the maximum time on any input of size  $n$

- **Average-case running time**

$T(n)$  : the expected time over all inputs of size  $n$

*\* need an assumption of statistical distribution of inputs (e.g. the uniform distribution)*

- **Best-case running time**

$T(n)$  : the minimum time on any input of size  $n$

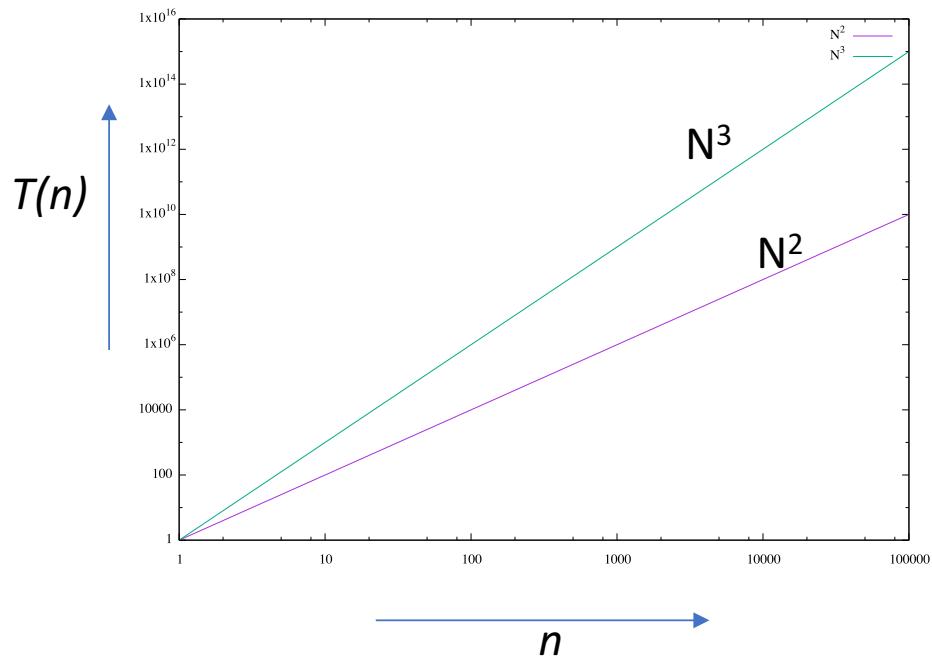
*\* we are not so interested in the best-case because a slow algorithm might have a very good best-case running time*

# What is Insertion Sort's Running Time?

- Running time of an algorithm depends on many factors: computer hardware, compiler, operating systems, ...
- BIG IDEA: Asymptotic Analysis
  - Ignore machine dependent constants
  - Look at the *rate of the growth of  $T(n)$  as  $n \rightarrow \infty$*

# Asymptotic Analysis : Theta-notation

$\Theta(g(n)) = \{f(n) : \text{there exist positive constants } c_1, c_2, \text{ and } n_0 \text{ such that}$   
 $0 \leq c_1g(n) \leq f(n) \leq c_2g(n) \text{ for all } n \geq n_0\}$ .<sup>1</sup>



$$(3n^3 + 2n + 2) = \theta(n^3)$$

$$(4n^2 + 3n + 4) = \theta(n^2)$$

*When  $n$  gets large enough,  $\theta(n^2)$  algorithm always beat  $\theta(n^3)$  algorithm*

# What is Insertion Sort's Running Time?

INSERTION-SORT ( $A$ )

```
1  for  $j = 2$  to  $A.length$ 
2    key =  $A[j]$ 
3    // Insert  $A[j]$  into the sorted
       sequence  $A[1..j - 1]$ .
4     $i = j - 1$ 
5    while  $i > 0$  and  $A[i] > key$ 
6       $A[i + 1] = A[i]$ 
7       $i = i - 1$ 
8     $A[i + 1] = key$ 
```

	cost	times
$c_1$	$n$	
$c_2$	$n - 1$	
$c_3$	0	$n - 1$
$c_4$	$n - 1$	
$c_5$	$\sum_{j=2}^n t_j$	
$c_6$	$\sum_{j=2}^n (t_j - 1)$	
$c_7$	$\sum_{j=2}^n (t_j - 1)$	
$c_8$	$n - 1$	

- Worst-case running time  
(reverse sorted input):

$$T(n) = \sum_{j=2}^N \theta(j) = \theta(N^2)$$

- Average-case running time  
(all ordering is equally likely):

$$T(n) = \sum_{j=2}^N \theta(j/2) = \theta(N^2)$$

$$\begin{aligned} T(n) &= c_1 n + c_2(n - 1) + c_4(n - 1) + c_5 \sum_{j=2}^n t_j + c_6 \sum_{j=2}^n (t_j - 1) \\ &\quad + c_7 \sum_{j=2}^n (t_j - 1) + c_8(n - 1). \end{aligned}$$

- Best-case running time  
(sorted input):

$$T(n) = \sum_{j=2}^N \theta(j) = \theta(N)$$

# Is Insertion-sort a fast algorithm?

- Moderately fast for small input size  $n$
- Not at all for large input size  $n$

# Pop-Quiz

- Illustrate the operation of insertion-sort on the array

$A = \langle 3, 41, 52, 26, 38, 57, 9, 49 \rangle$

# Summary: Insertion Sort

- Worst case running time is  $\theta(N^2)$
- An excellent sorting method for partially sorted arrays
- Moderately fast for small arrays
- Sort “in place”
- Very slow for large arrays

# Divide-and-Conquer

- **Divide:** the problem into a number of sub-problems that are smaller instances of the same problem.
- **Conquer:** the sub-problems by solving them recursively. If the sub-problem sizes are small enough, however, just solve the sub-problems in a straightforward manner.
- **Combine:** the solutions to the sub-problems into the solution for the original problem.

# Mergesort

- **Divide:** Divide the  $n$ -element sequence to be sorted into two subsequences of  $n/2$  elements each.
- **Conquer:** Sort the two subsequences recursively using merge sort.
- **Combine:** *Merge* the two sorted subsequences to produce the sorted answer.

# Mergesort

8	2	4	9	3	6	5	1
---	---	---	---	---	---	---	---

# Mergesort

8	2	4	9	3	6	5	1
---	---	---	---	---	---	---	---

8	2	4	9
---	---	---	---

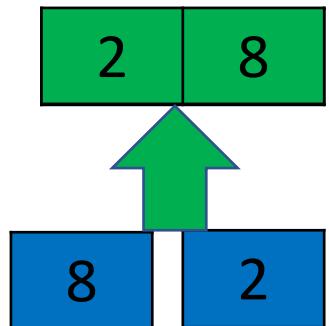
# Mergesort

8	2	4	9	3	6	5	1
---	---	---	---	---	---	---	---

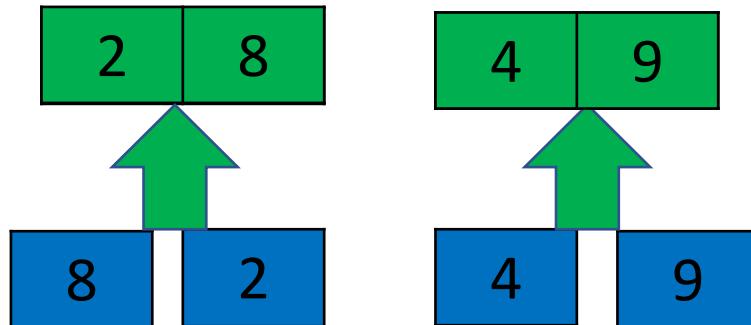
8	2	4	9
---	---	---	---

8	2
---	---

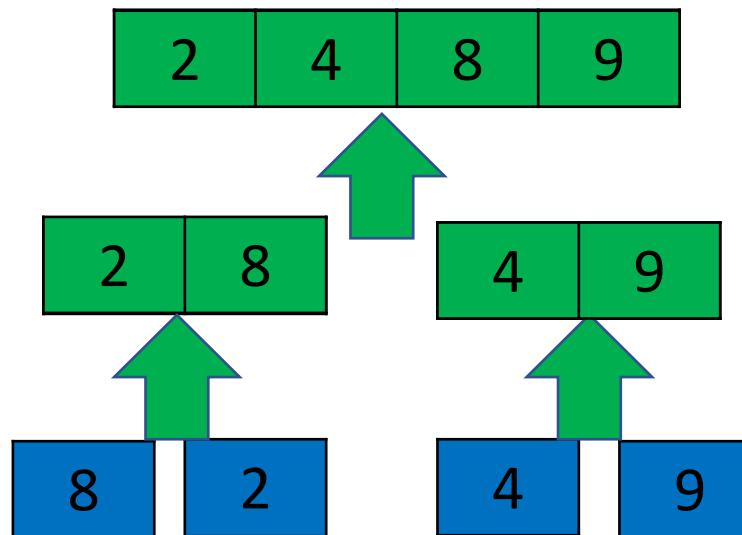
# Mergesort



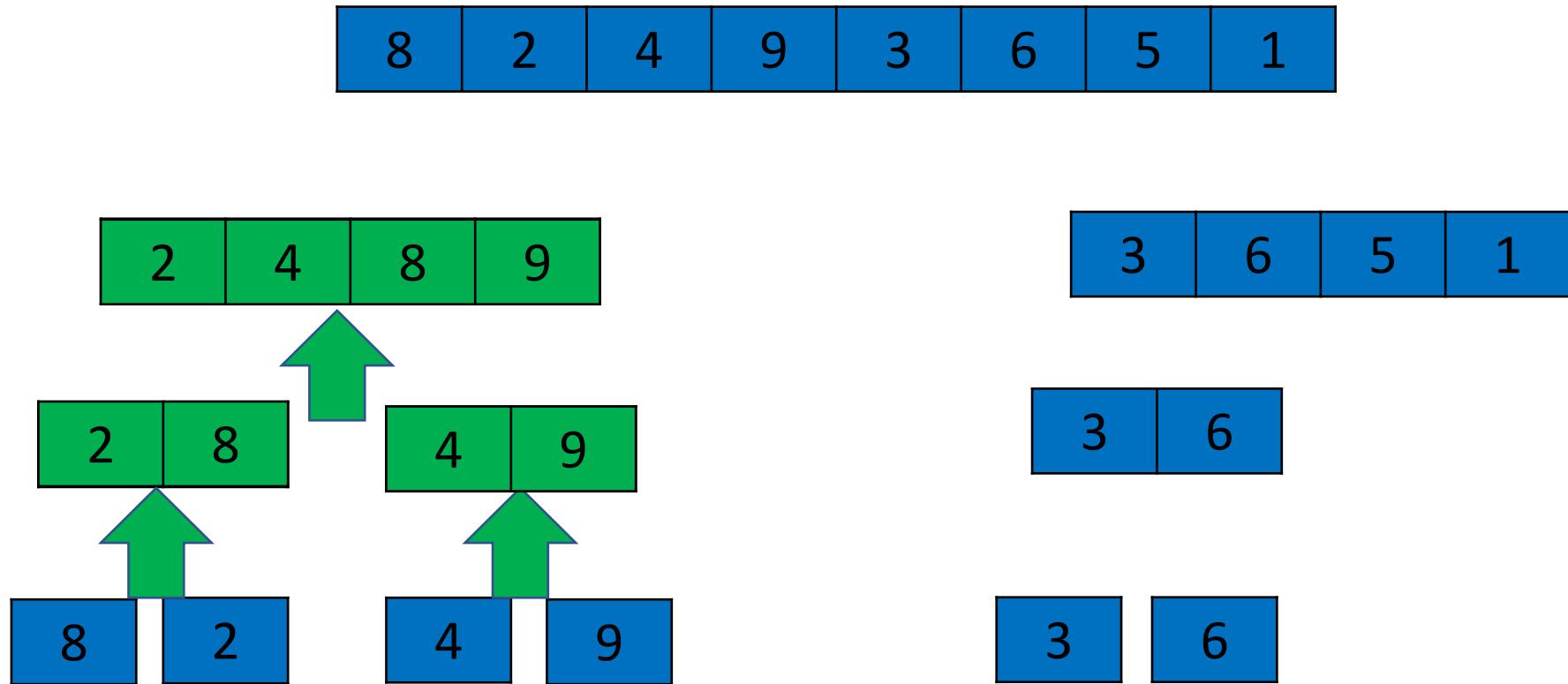
# Mergesort



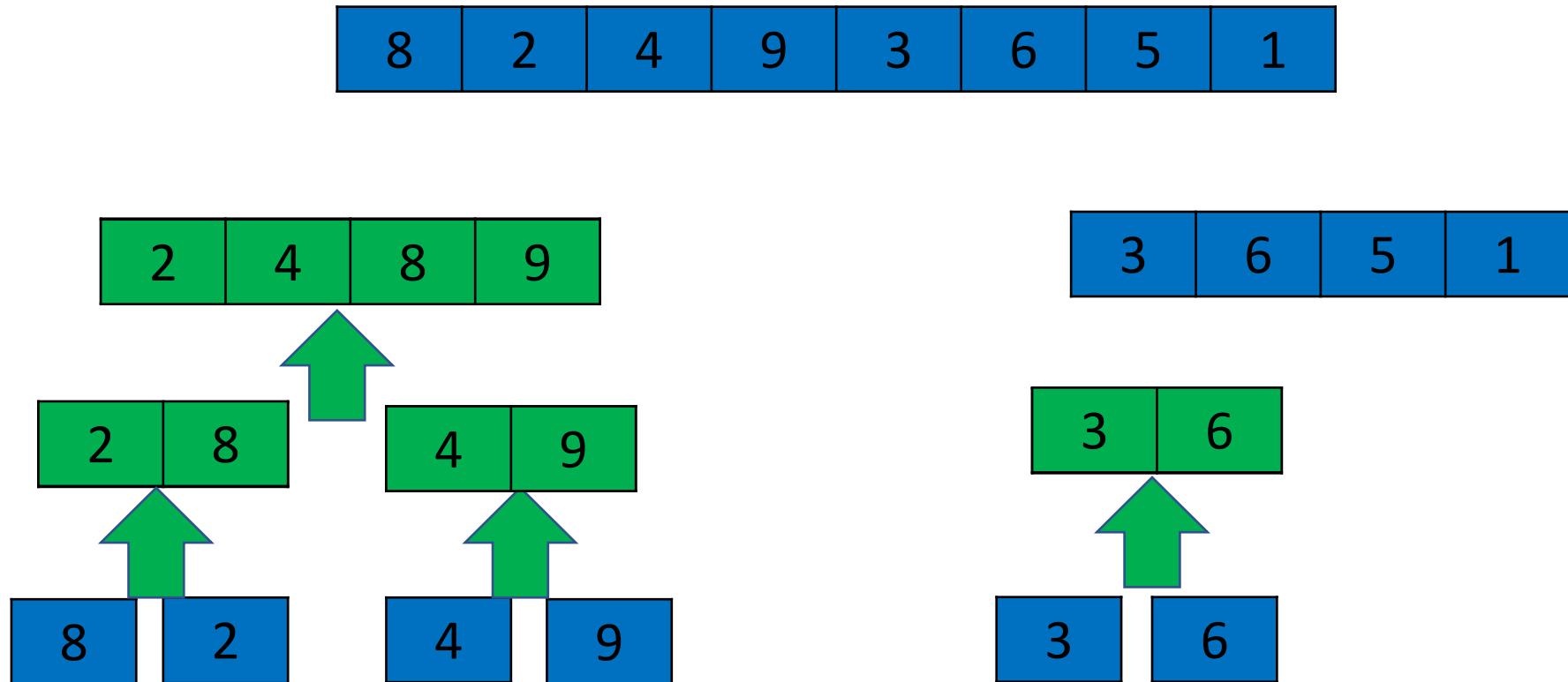
# Mergesort



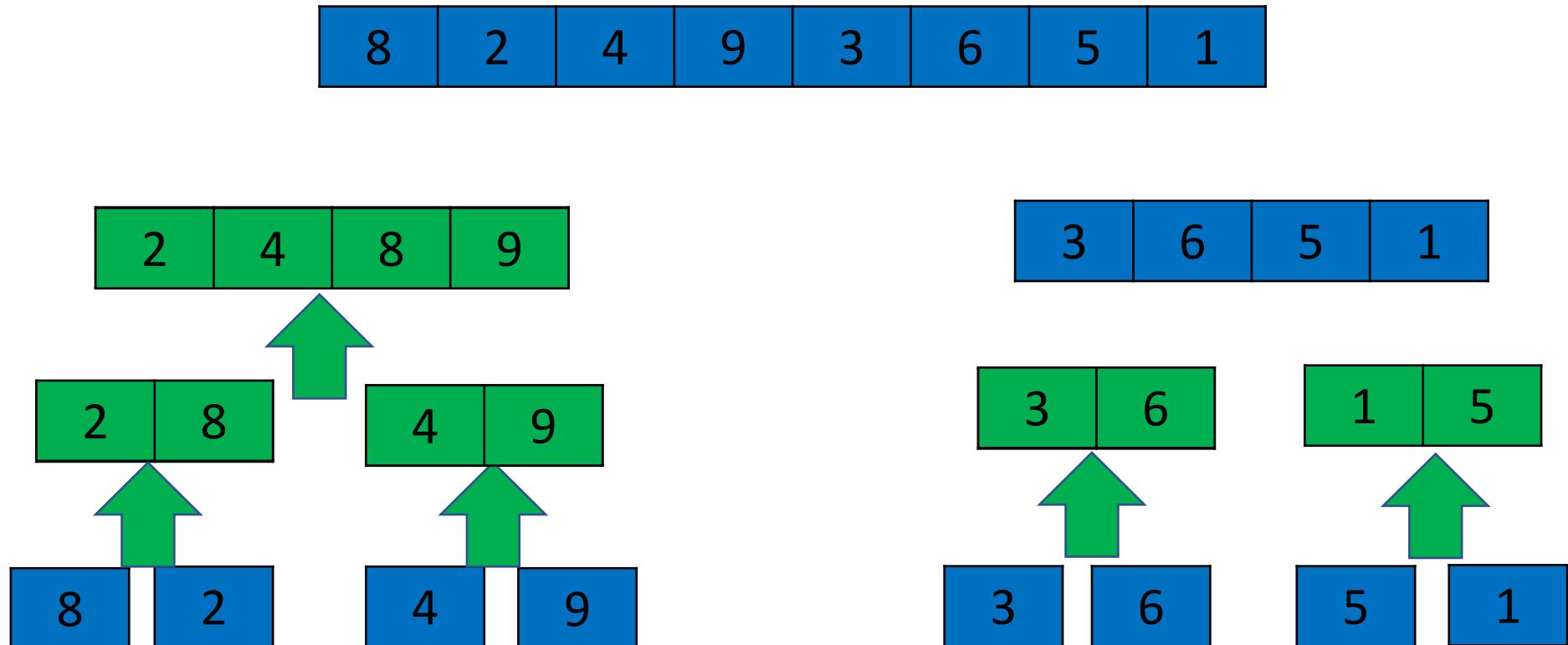
# Mergesort



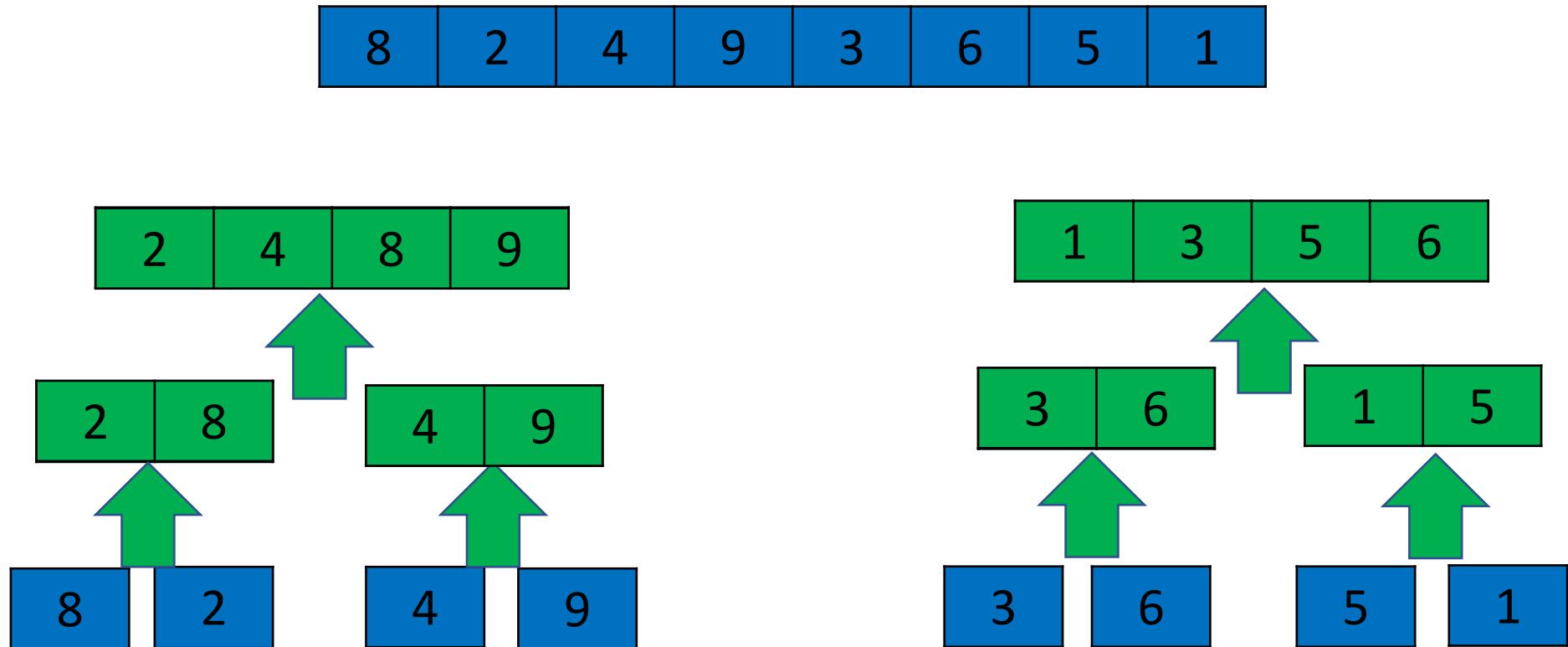
# Mergesort



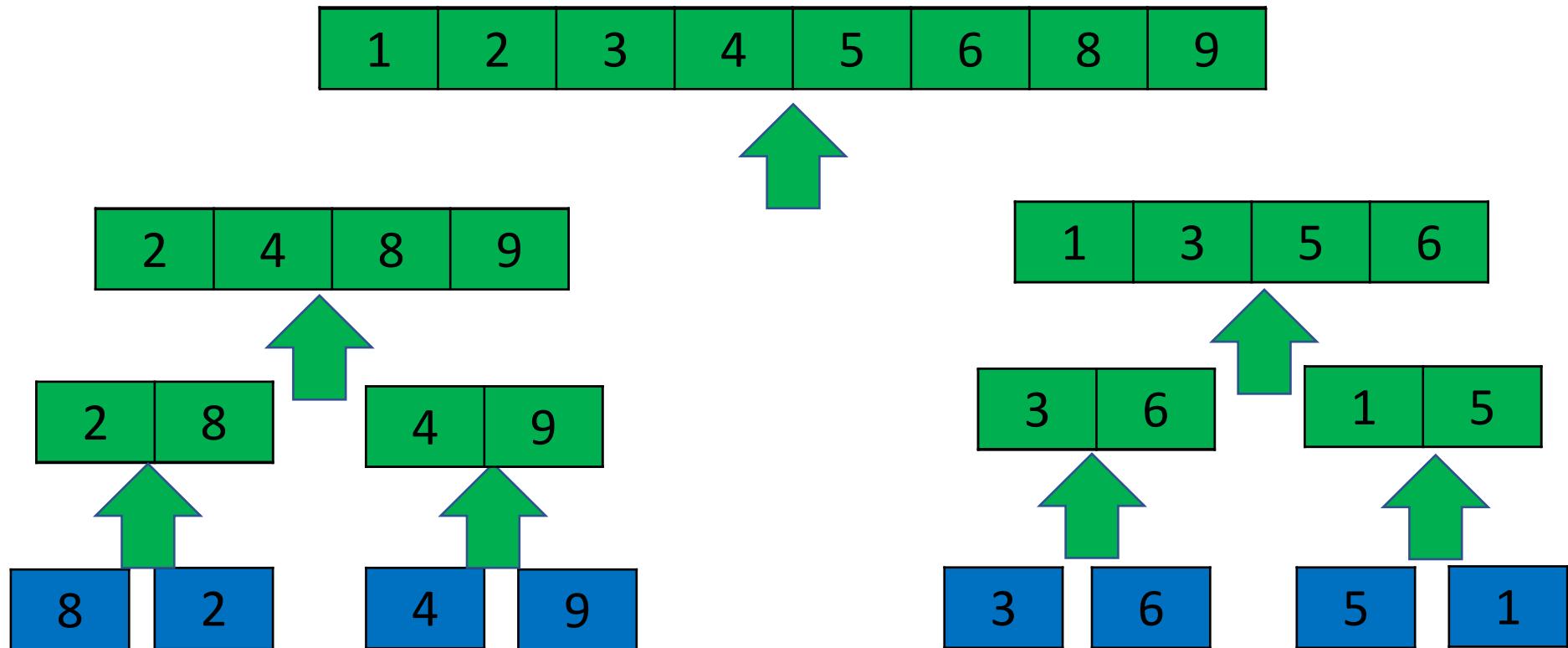
# Mergesort



# Mergesort



# Mergesort

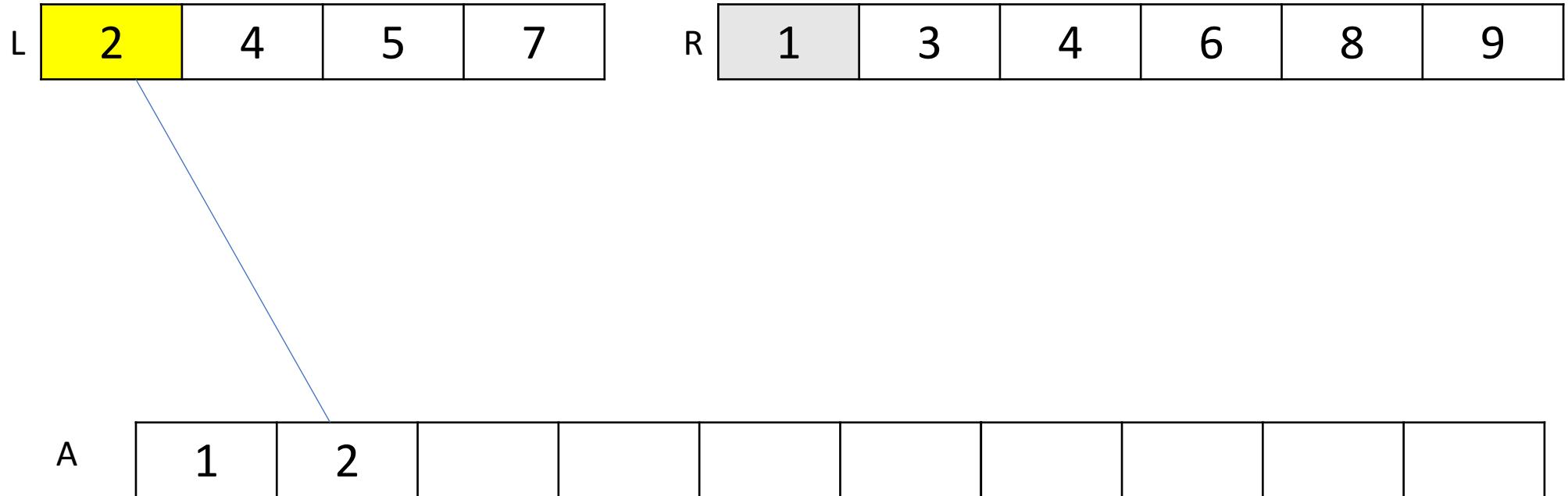


# Merge two sorted arrays...

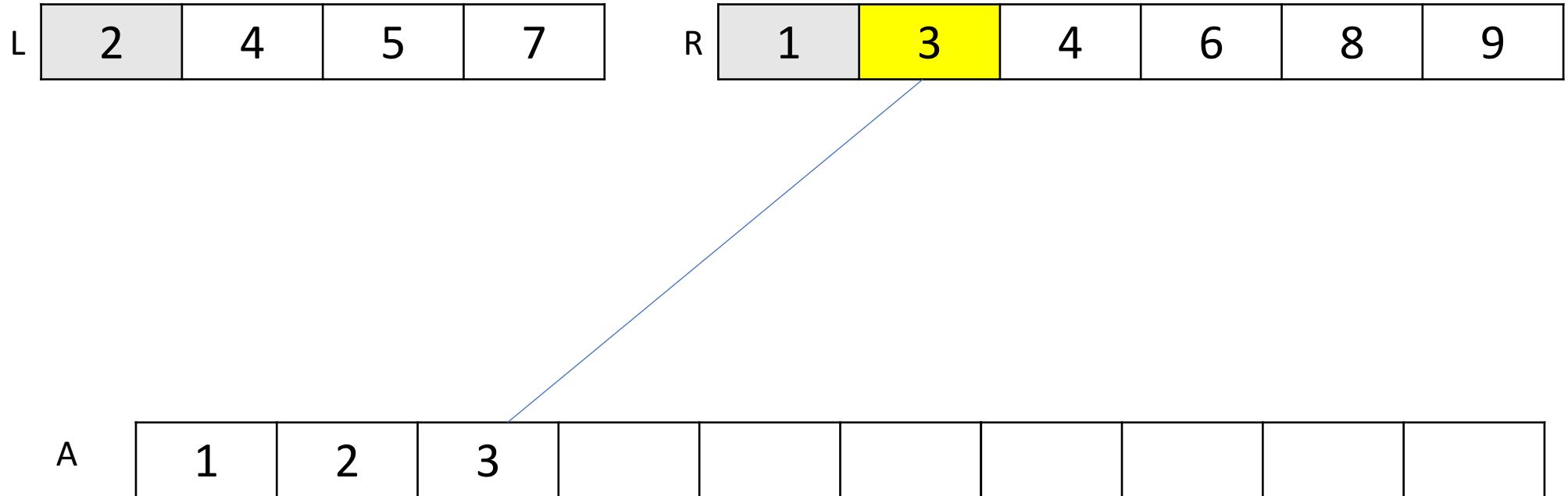
L	2	4	5	7
---	---	---	---	---

R	1	3	4	6	8	9
---	---	---	---	---	---	---

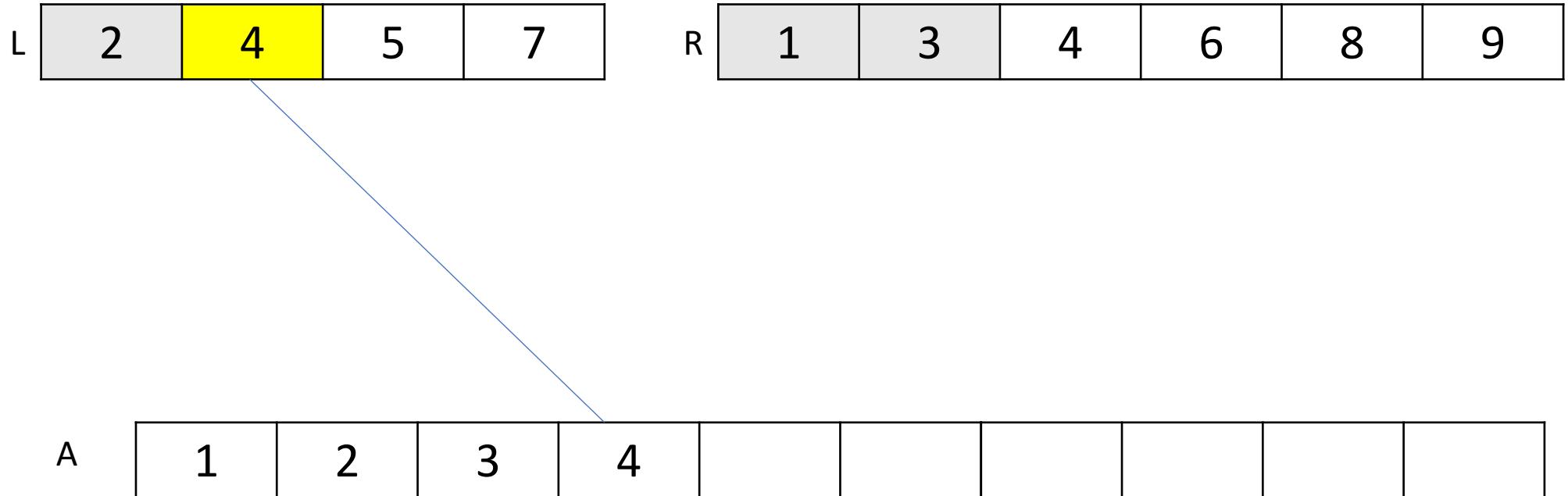
# Merge two sorted arrays...



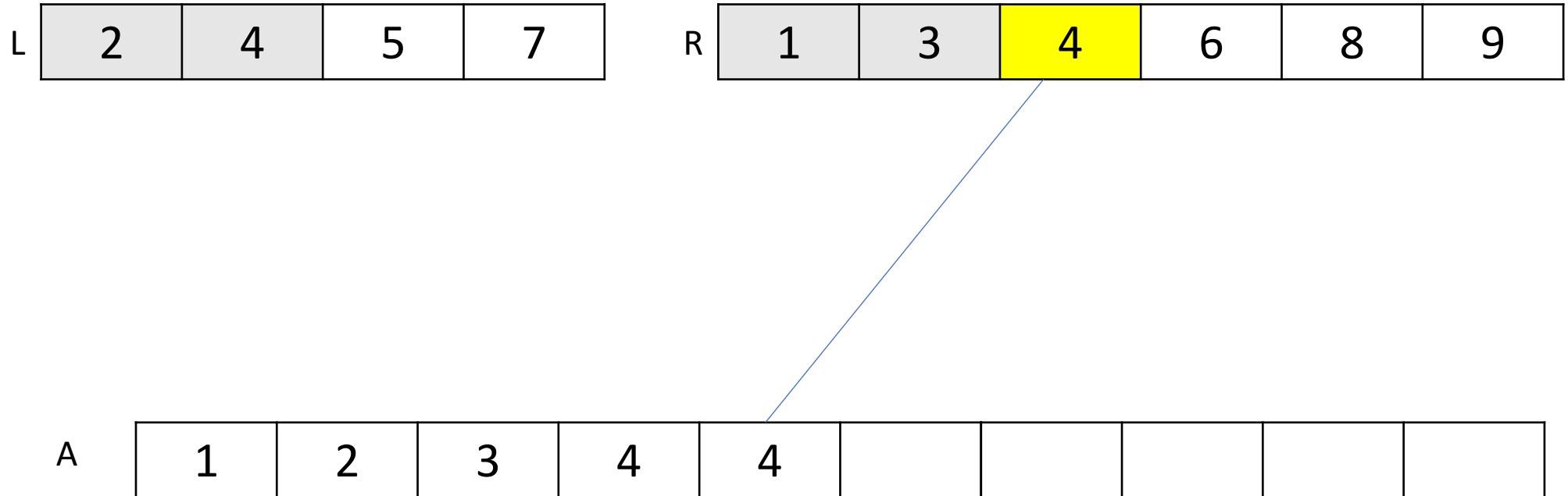
# Merge two sorted arrays...



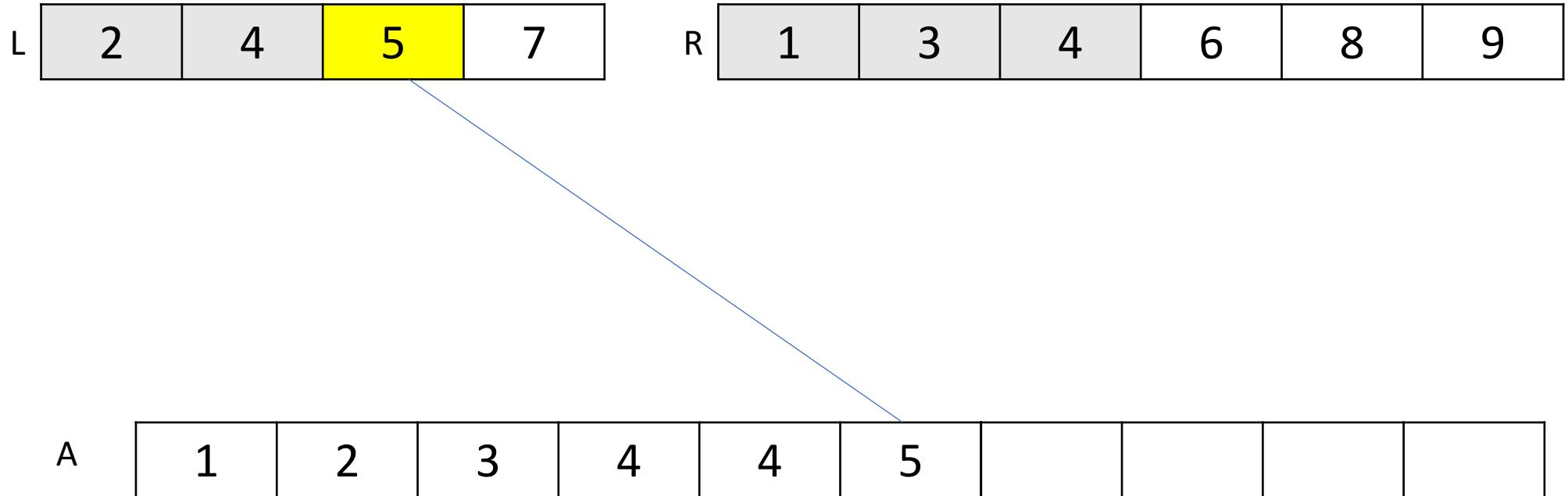
# Merge two sorted arrays...



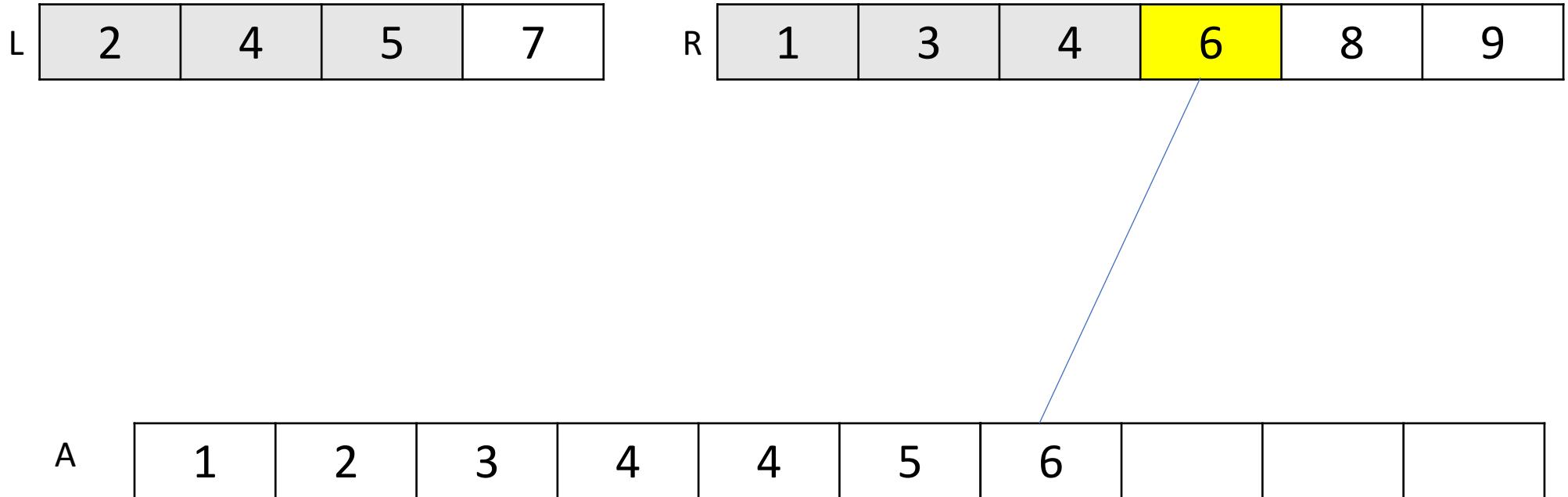
# Merge two sorted arrays...



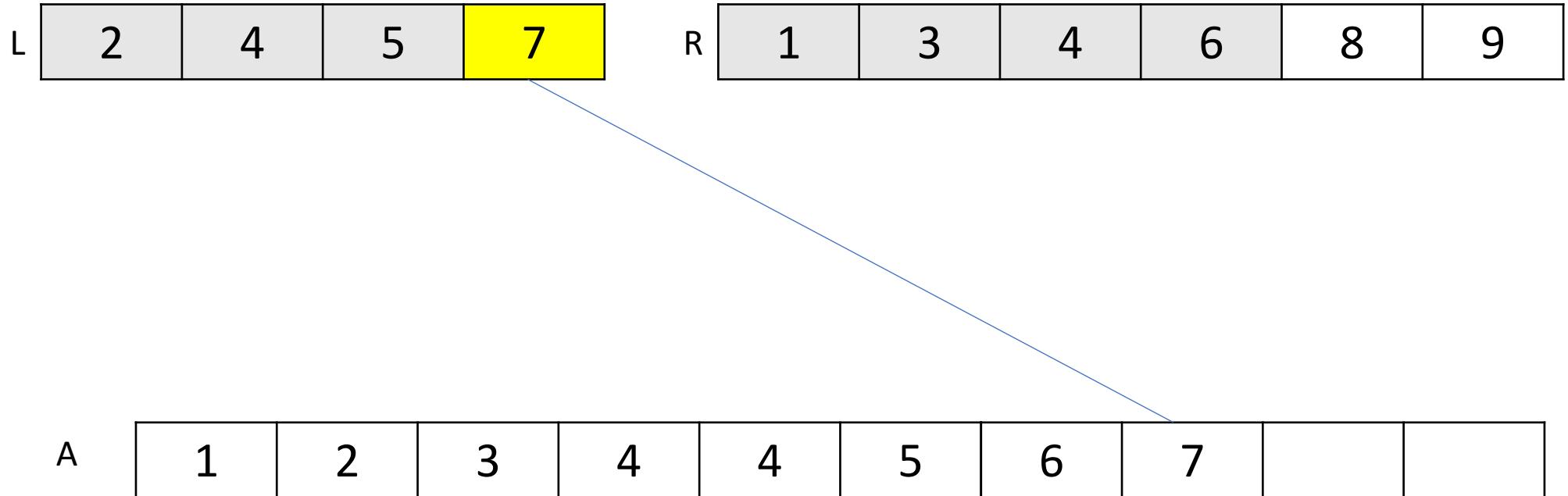
# Merge two sorted arrays...



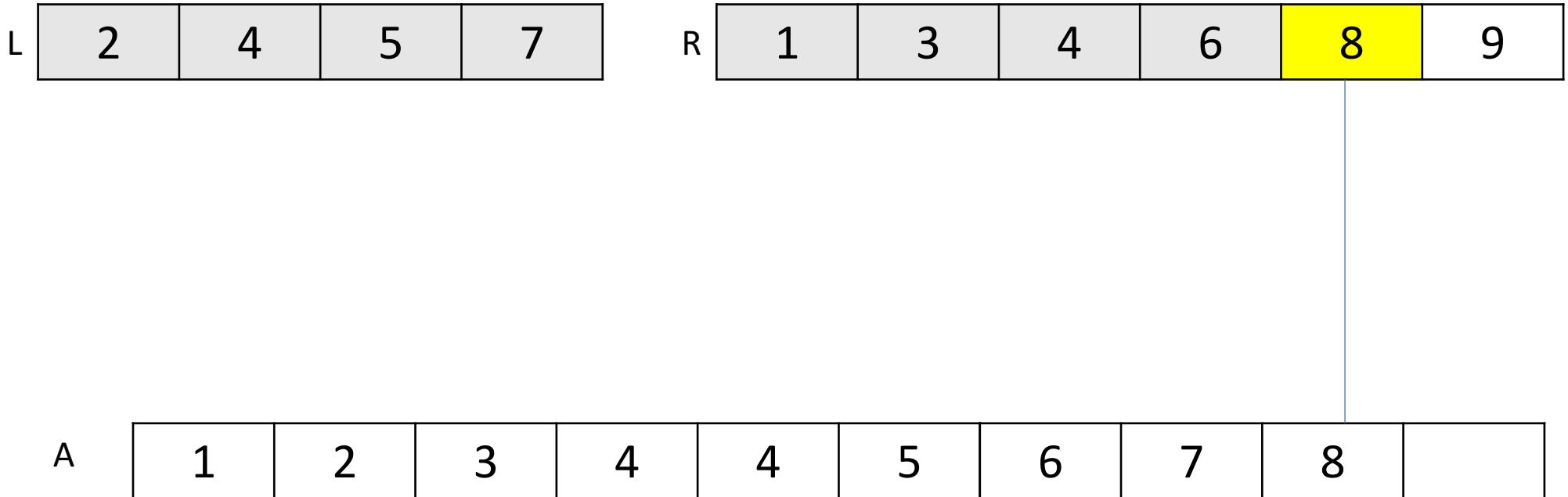
# Merge two sorted arrays...



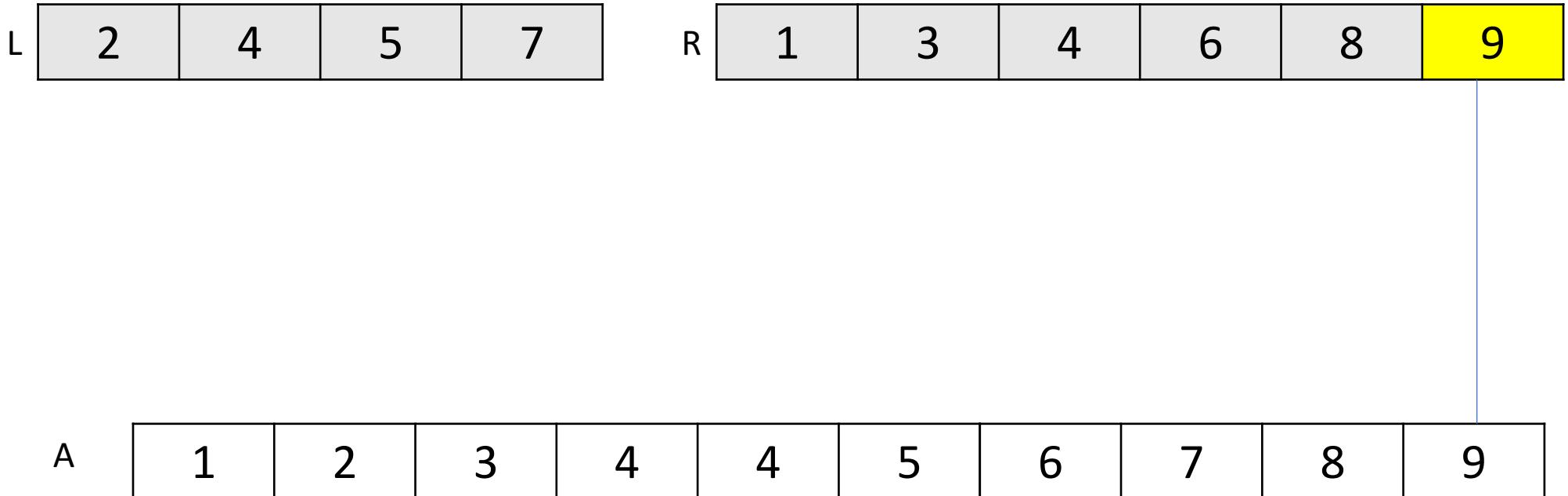
# Merge two sorted arrays...



# Merge two sorted arrays...



# Merge two sorted arrays...



# Merge A[p..q] with A[q+1..r] into A[p..r]

MERGE( $A, p, q, r$ )

```
1   $n_1 = q - p + 1$ 
2   $n_2 = r - q$ 
3  let  $L[1..n_1 + 1]$  and  $R[1..n_2 + 1]$  be new arrays
4  for  $i = 1$  to  $n_1$ 
5       $L[i] = A[p + i - 1]$ 
6  for  $j = 1$  to  $n_2$ 
7       $R[j] = A[q + j]$ 
8   $L[n_1 + 1] = \infty$ 
9   $R[n_2 + 1] = \infty$ 
10  $i = 1$ 
11  $j = 1$ 
12 for  $k = p$  to  $r$ 
13     if  $L[i] \leq R[j]$ 
14          $A[k] = L[i]$ 
15          $i = i + 1$ 
16     else  $A[k] = R[j]$ 
17          $j = j + 1$ 
```

*Merge takes  $\theta(N)$  to merge  $N$  elements*

# Mergesort

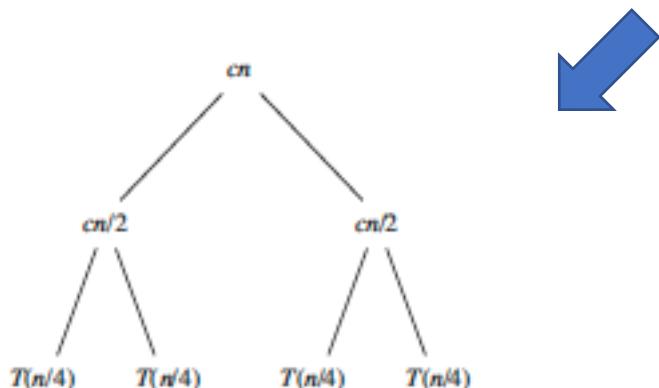
```
MERGE-SORT( $A, p, r$ )
```

```
1  if  $p < r$  .....  $\theta(1)$ 
2     $q = \lfloor (p + r)/2 \rfloor$ 
3    MERGE-SORT( $A, p, q$ )
4    MERGE-SORT( $A, q + 1, r$ )
5    MERGE( $A, p, q, r$ ) .....  $\theta(N)$ 
```

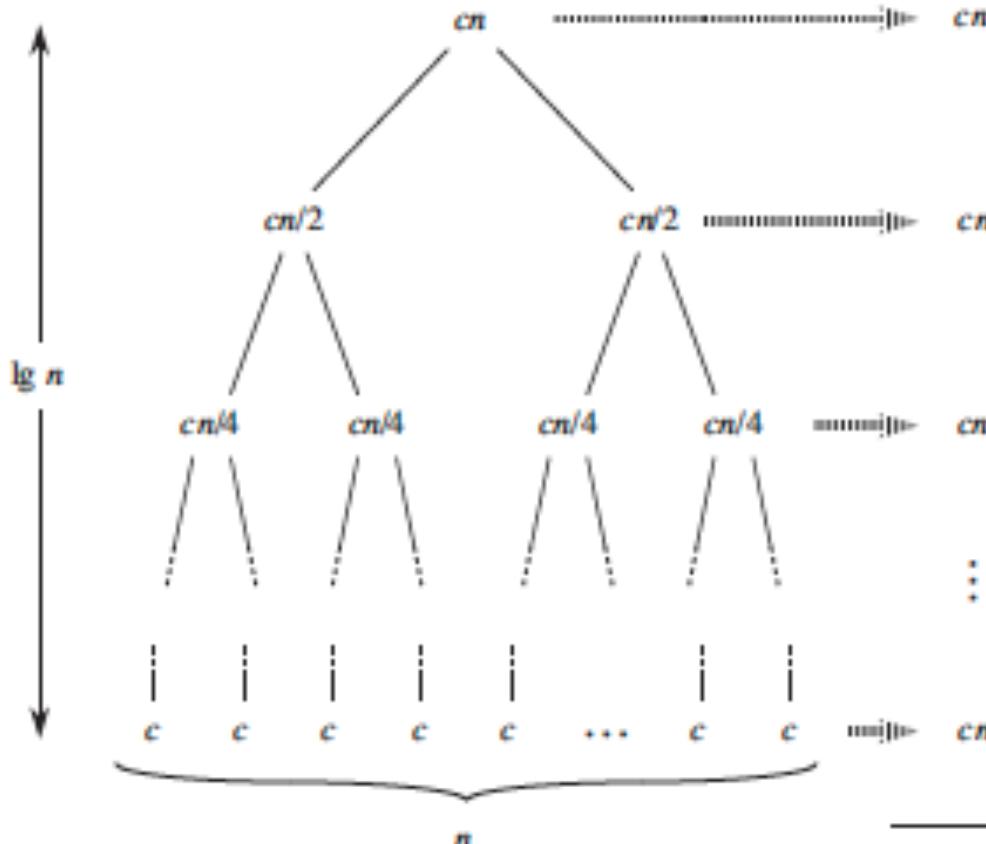
## Running time of Mergesort

$$T(n) = \begin{cases} \Theta(1) & \text{if } n = 1; \\ 2T(n/2) + \Theta(n) & \text{if } n > 1. \end{cases}$$

# Using Recursion Tree to solve for T(n)



# Using Recursion Tree to solve for $T(n)$



Total:  $cn \lg n + cn$

\* Mergesort running time is **linearithmic**.  $T(N) = \theta(N \lg N)$

# Pop-Quiz

- Illustrate the operation of merge-sort on the array

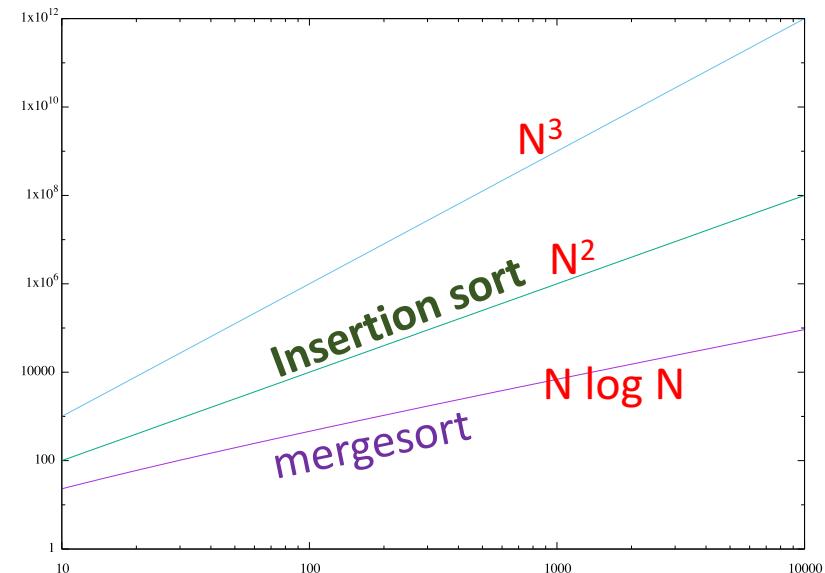
$$A = \langle 3, 41, 52, 26, 38, 57, 9, 49 \rangle$$

- Compare running time for different size of n:

n	$T(n) = 2n^2$	$T(n) = 8n \lg n$
2		
4		
8		
16		
32		
64		
128		

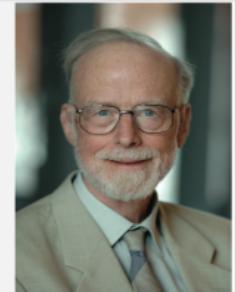
# Summary: MergeSort

- Worst case running time is  $\theta(N \lg N)$
- For large enough input size, mergesort always beat insertion sort.
- NOT sort “in place”
- Mergesort is **not optimal with respect to space usage**



# Quicksort

- Proposed in 1961 by C.A.R. Hoare
- Divide-and-Conquer algorithm
- Sorts “in place” (like insertion sort)
- Highly practical (with appropriate tuning)



Tony Hoare  
1980 Turing Award



ALGORITHM 64  
QUICKSORT  
C. A. R. HOARE  
Elliott Brothers Ltd., Borehamwood, Hertfordshire, Eng.

```
procedure quicksort (A,M,N); value M,N;
    array A; integer M,N;
comment Quicksort is a very fast and convenient method of
sorting an array in the random-access store of a computer. The
entire contents of the store may be sorted, since no extra space is
required. The average number of comparisons made is  $2(M-N) \ln$ 
 $(N-M)$ , and the average number of exchanges is one sixth this
amount. Suitable refinements of this method will be desirable for
its implementation on any actual computer;
begin      integer I,J;
    if M < N then begin partition (A,M,N,I,J);
                    quicksort (A,M,J);
                    quicksort (A, I, N)
    end
end      quicksort
```

Communications of the ACM (July 1961)

# Quicksort

- **Divide:** Partition the array into two subarrays around a *pivot*  $x$  such that elements in lower subarray  $\leq x \leq$  elements in upper subarray



- **Conquer:** Sort the two subsequences recursively using quicksort.
- **Combine:** -

# Quicksort

```
QUICKSORT( $A, p, r$ )
```

```
1  if  $p < r$ 
2       $q = \text{PARTITION}(A, p, r)$ 
3       $\text{QUICKSORT}(A, p, q - 1)$ 
4       $\text{QUICKSORT}(A, q + 1, r)$ 
```

```
PARTITION( $A, p, r$ )
```

```
1   $x = A[r]$ 
2   $i = p - 1$ 
3  for  $j = p$  to  $r - 1$ 
4      if  $A[j] \leq x$ 
5           $i = i + 1$ 
6          exchange  $A[i]$  with  $A[j]$ 
7  exchange  $A[i + 1]$  with  $A[r]$ 
8  return  $i + 1$ 
```

# quicksort(A, 1, 8)

$i=0 \quad j=1$

$x = A[8]$



$\text{partition}(A, p=1, r=8)$

$i=1 \quad j=2$

$x = A[8]$



$i=2 \quad j=3$

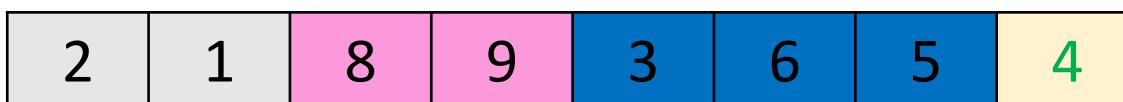
$x = A[8]$



$i=2$

$j=4$

$x = A[8]$



$i=3$

$j=5$

$x = A[8]$



# quicksort(A, 1, 8)

$i=3$	$j=6$	$x = A[8]$					
2	1	3	9	8	6	5	4

$\text{partition}(A, p=1, r=8)$

$i=3$	$j=7$	$x = A[8]$					
2	1	3	9	8	6	5	4

$i=3$	$j=7$	$x = A[8]$					
2	1	3	4	8	6	5	9

$\text{return } 4$

$\text{quicksort}(A, p=1, r=3)$

$\text{quicksort}(A, p=5, r=8)$

# quicksort(A, 1, 3)

$i=0$

$j=1$

$x = A[3]$

2	1	3	4	8	6	5	9
---	---	---	---	---	---	---	---

$\text{partition}(A, p=1, r=3)$

$i=1$

$j=1$

$x = A[3]$

2	1	3	4	8	6	5	9
---	---	---	---	---	---	---	---

$i=2$

$j=2$

$x = A[3]$

2	1	3	4	8	6	5	9
---	---	---	---	---	---	---	---

$\text{return } 3$

quicksort(A, 1, 3)

quicksort(A, 1, 8)

# quicksort(A, 1, 2)

$i=0 \quad j=1 \quad x = A[2]$

2	1	3	4	8	6	5	9
---	---	---	---	---	---	---	---

$\text{partition}(A, p=1, r=2)$

$i=0 \quad j=2 \quad x = A[2]$

2	1	3	4	8	6	5	9
---	---	---	---	---	---	---	---

$i=0 \quad j=2 \quad x = A[2]$

1	2	3	4	8	6	5	9
---	---	---	---	---	---	---	---

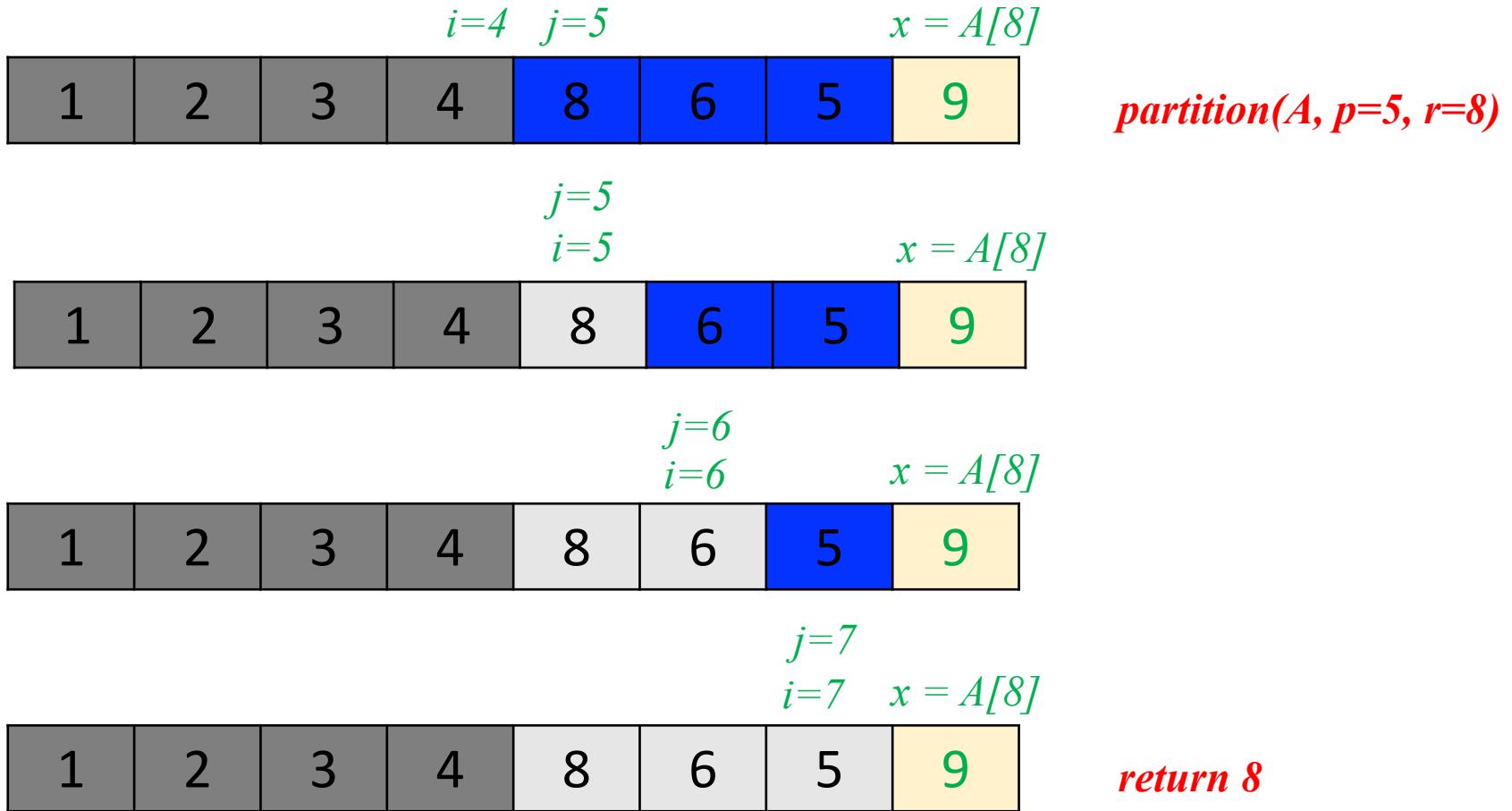
$\text{return } 1$

quicksort(A, 1, 2)

quicksort(A, 1, 3)

quicksort(A, 1, 8)

# quicksort(A, 5, 8)



quicksort(A, 5, 8)

quicksort(A, 1, 8)

# quicksort(A, 5, 7)

$$i=4 \quad j=5 \quad x = A[7]$$

1	2	3	4	8	6	5	9
---	---	---	---	---	---	---	---

*partition(A, 5, 7)*

$$i=4 \quad j=6 \quad x = A[7]$$

1	2	3	4	8	6	5	9
---	---	---	---	---	---	---	---

$$i=4 \quad j=6 \quad x = A[7]$$

1	2	3	4	8	6	5	9
---	---	---	---	---	---	---	---

$$i=4 \quad j=6 \quad x = A[7]$$

1	2	3	4	5	6	8	9
---	---	---	---	---	---	---	---

*return 5*

quicksort(A, 5, 7)

quicksort(A, 5, 8)

quicksort(A, 1, 8)

quicksort(A, 6, 7)

$i=5 \quad j=6 \quad x = A[7]$



$\text{partition}(A, 6, 7)$

$i=6$   
 $j=6 \quad x = A[7]$



$\text{return } 7$

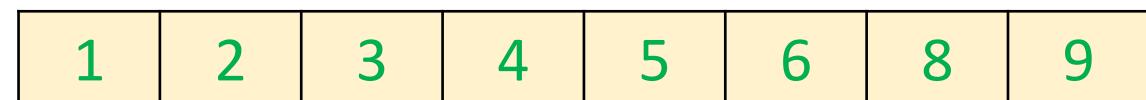
quicksort(A, 6, 7)

quicksort(A, 5, 7)

quicksort(A, 5, 8)

quicksort(A, 1, 8)

Sorted...



# Worst Case Running Time of Quicksort

- Input already sorted or reverse sorted
- Partition around min/max

$$\begin{aligned}T(n) &= T(0) + T(n-1) + \theta(N) \\&= \theta(1) + T(n-1) + \theta(N) \\&= T(n-1) + \theta(N) \\&= \theta(N^2)\end{aligned}$$

# Best Case Running Time of Quicksort

- Partition splits the array evenly

$$\begin{aligned} T(n) &= 2T(n/2) + \theta(N) \\ &= \theta(N \lg N) \quad \dots\dots \text{ same as mergesort} \end{aligned}$$

*How to make sure that Partition always splits the array evenly ?*

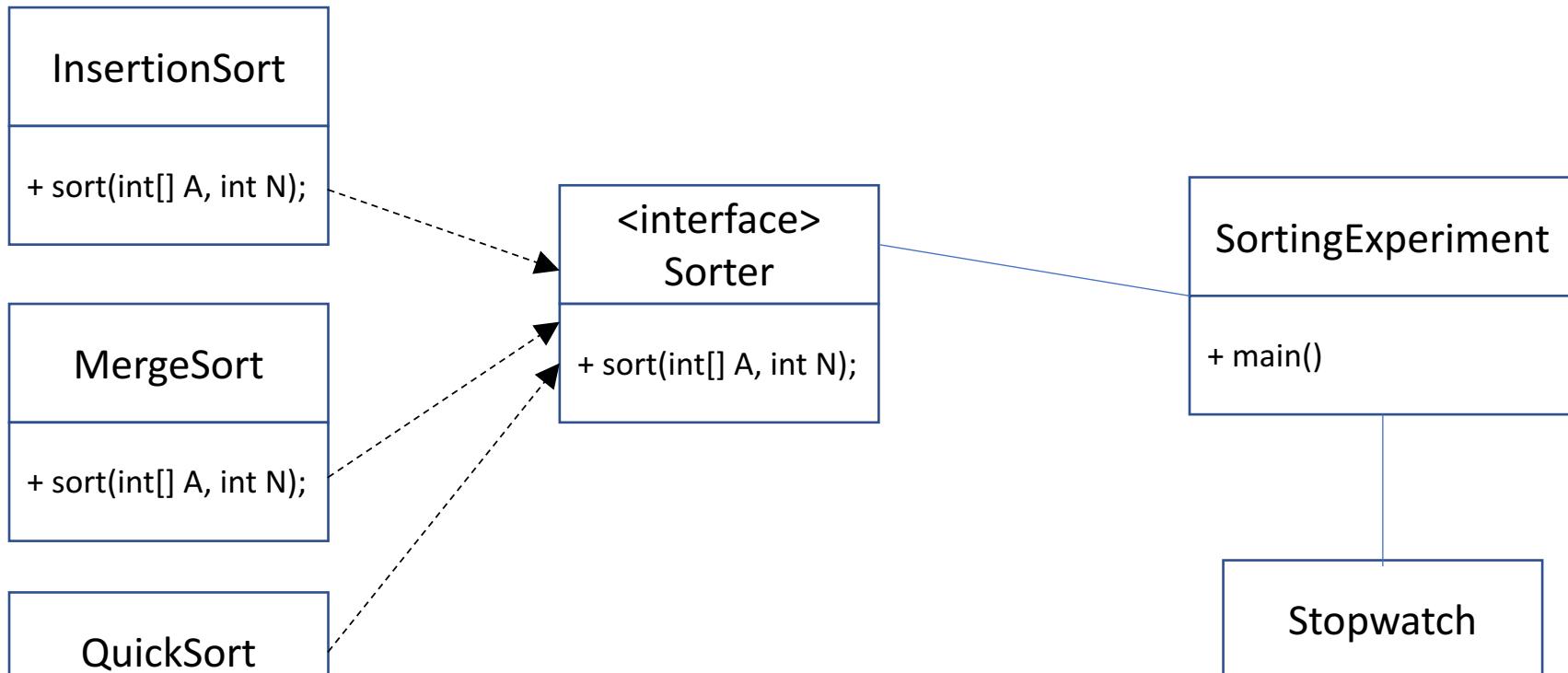
# Randomized Quicksort

- **Idea:** Partition the array around a *random* element
- Running time is independent of the order of input
- The worst-case is determined only by the output of a random number generator

# Summary: Quicksort

- A great general purpose sorting algorithm
- Sort “in place”
- Typically over twice as fast as mergesort
- Performance can be substantially improved by various code tuning

# Java Implementation



....

```

public Stopwatch() {
    start = System.nanoTime();
}
public long elapsed() {
    return (System.nanoTime() - start);
}

/****** Measure Sorting Time *****/
w = new Stopwatch();
sorter.sort(A, N);
totTime += w.elapsed();

***** */

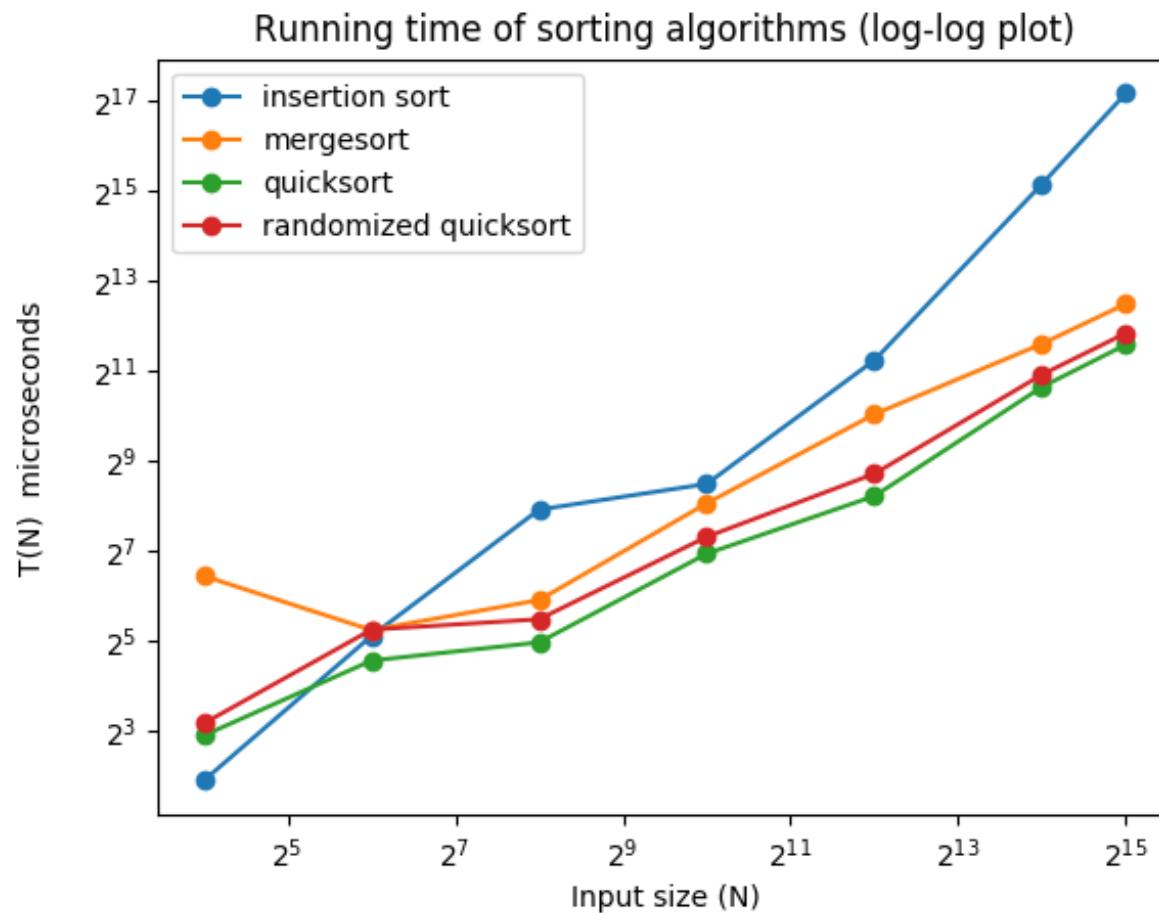
```

kulwadee-mbair:int320 ann\$ javac dsa/sorting/*.java							
kulwadee-mbair:int320 ann\$ java -cp . dsa.sorting.SortingExperiment							
#Size: 16 64 256 1024 4096 16384 32768							
#Time unit: microseconds							
insertion sort:	8.21	69.15	363.09	460.71	2513.76	36552.40	146835.95
mergesort:	209.65	50.06	91.96	264.72	1105.28	3217.84	5638.38
quicksort:	7.82	19.89	27.29	116.54	325.38	1432.15	3152.02
randomized quicksort:	10.23	40.85	44.50	143.15	588.93	1773.45	3640.32

```
java -cp . dsa.sorting.SortingExperiment > sorting.txt
```

```
python plot_sorting.py sorting.txt
```

```
# output a log-log plot of the running time to `sorting.png`
```



# Pop-Quiz

- Given specifications of two computers *A* and *B*:

**Computer A:** can execute *2 billion instructions per second*.

**Computer B:** can execute *10 million instructions per second*.

Answer the following questions.

- Computer *A* is \_\_\_\_\_ times (*faster / slower*) than Computer *B*.
- Suppose that a program for Computer *A* that implements an *insertion sort* algorithm requires  $4n^2$  **instructions to sort *n* numbers**. How long does it take for Computer *A* to sort **10 million numbers using this insertion sort** program?
- Suppose that a program for Computer *B* that implements a *merge sort* algorithm requires  $100n \log n$  **instructions to sort *n* numbers**. How long does it take for Computer *B* to sort **10 million numbers using this merge sort** program?